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## Paper: **Energy-Efficient Subcarrier Assignment and Power Allocation in OFDMA Systems With Max-Min Fairness Guarantees**

-Y. Li, M. Sheng, C. W. Tan, Y. Zhang, Y. Sun, X. Wang, Y. Shi, J. Li

### Aim:

- Guarantee individual link fairness: max-min energy efficiency-optimal problem (MEP).
- Maximize the energy efficiency subject to **rate requirements, transmit power, subcarrier assignment** constraints.

### Setup:

- Single cell uplink OFDMA,  $K$  active users,  $N$  subcarriers.
- Define  $\mathbf{P} = (P_{k,n})$ ,  $P_{k,n}$ : Tx power of link  $k$  on subcarrier  $n$ .
- Tx rate,  $r_{k,n} = \log_2 \left( 1 + \frac{P_{k,n}|h_{k,n}|^2}{N_0 B/N} \right)$ .
- Total Tx rate on link  $k$ ,  $R_k(\boldsymbol{\rho}, \mathbf{P}) = \sum_{n=1}^N \rho_{k,n} r_{k,n}$ ;  $\rho_{k,n}$ : Indicator variable.
- Total Tx power on link  $k$ ,  $P_k(\boldsymbol{\rho}, \mathbf{P}) = \sum_{n=1}^N \rho_{k,n} P_{k,n}$ .
- $R_{\text{tot}}(\boldsymbol{\rho}, \mathbf{P}) = \sum_{k=1}^K R_k(\boldsymbol{\rho}, \mathbf{P})$ ,  $P_{\text{tot}}(\boldsymbol{\rho}, \mathbf{P}) = \sum_{k=1}^K P_k(\boldsymbol{\rho}, \mathbf{P})$ .
- Energy efficiency:  $\eta_k^{\text{EE}} = R_k(\boldsymbol{\rho}, \mathbf{P})/P_k(\boldsymbol{\rho}, \mathbf{P})$ , and  $\eta^{\text{EE}} = R_{\text{tot}}(\boldsymbol{\rho}, \mathbf{P})/P_{\text{tot}}(\boldsymbol{\rho}, \mathbf{P})$ .

Problem  
formulation:

$$\begin{aligned} & \max_{\rho, \mathbf{P}} \min_k \eta_k^{\text{EE}} \\ \text{C1: } & \sum_{n=1}^N \rho_{k,n} r_{k,n} \geq R_{\text{req}}, \forall k \\ \text{C2: } & \sum_{k=1}^K \rho_{k,n} r_{k,n} \leq 1, \forall n \\ \text{C3: } & \sum_{n=1}^N \rho_{k,n} P_{k,n} \leq P_k^{\text{max}}, \forall k \\ \text{C4: } & P_{k,n} \geq 0, \forall k, n \\ \text{C5: } & \rho_{k,n} \in \{0, 1\}, \forall k, n \end{aligned}$$

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**Algorithm 1** Iterative Subcarrier Assignment and Power Allocation Algorithm

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1: **Initialization**

- Set the maximum iteration number  $i_{\text{max}}$  and error tolerance threshold  $\varepsilon$ .
- Set iteration index  $i = 0$  and the maximum energy efficiency  $\eta^i = 0$ .

2: **repeat**

3: Solve (12) for the given  $\eta^i$  to obtain  $\{\rho^i, \mathbf{P}^i\}$ .

4: **if**  $\left| \min_k [R_k(\rho^i, \mathbf{P}^i) - \eta^i \text{PC}_k(\rho^i, \mathbf{P}^i)] \right| < \varepsilon$  **then**

5:  $\{\rho^{\text{opt}}, \mathbf{P}^{\text{opt}}\} = \{\rho^i, \mathbf{P}^i\}$ .

6:  $\eta_{\text{EE}}^{\text{opt}} = \min_k \frac{R_k(\rho^i, \mathbf{P}^i)}{\text{PC}_k(\rho^i, \mathbf{P}^i)}$ .

7: **break**.

8: **else**

9: Set  $\eta^{i+1} = \min_k \frac{R_k(\rho^i, \mathbf{P}^i)}{\text{PC}_k(\rho^i, \mathbf{P}^i)}$ .

10:  $i = i + 1$ .

11: **end if**

12: **until**  $i > i_{\text{max}}$ .

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- Mixed-integer non-linear programming - hard to solve, non-convex even after relaxing  $\rho_{k,n} \in [0, 1]$ .
- **Replace objective by:**  $\max_{\rho, \mathbf{P}} \min_k R_k(\rho, \mathbf{P}) - \eta P_k(\rho, \mathbf{P})$ .

- Objective non-smooth.
- Introduce new variable  $\varphi$ , replace objective by:  $\max_{\rho, \mathbf{P}, \varphi} \varphi$ .
- **Add new constraint:**  $C6 : R_k(\rho, \mathbf{P}) - \eta P_k(\rho, \mathbf{P}) \geq \varphi$ .
- Apply Lagrangian dual decomposition on the new problem:  $L(\rho, \mathbf{P}, \varphi, \beta, \mu, \nu, \gamma)$ .

### Iterate:

- Solve for optimal  $P_{k,n}^*$  and  $\rho_{k,n}^*$  by differentiating the Lagrangian.
- Update Lagrange multipliers using subgradient projection.
- Separate subcarrier assignment and power allocation problems to devise simpler algorithms.

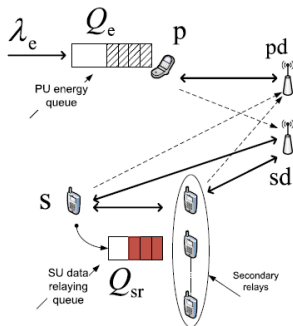
## Paper: **A Sparsity-Aware Cooperative Protocol for Cognitive Radio Networks With Energy-Harvesting Primary User.**

-A. El Shafie, N. Al-Dhahir, and R. Hamila.

Aim:

- Design a dynamic relaying cooperative protocol for a CR setting - SUs coexist with energy-harvesting PUs.
- CS-aided protocol where the SUs also act as dynamic relay nodes.
- Design of a beamformer so that relay nodes can transmit without interference to PU destination.

Setup:



- Each channel has 1 PU and 1 SU,  $N$  SUs act as relaying nodes.
- $h_{n_1, n_2}$ : channel coefficient from node  $n_1$  to  $n_2$ .

## Proposed Protocol:

- SU source & SU relays sense the channel for PU activity.
- SU source transmits packets probabilistically (prob. 1 if direct link is not in outage, and prob.  $\omega$  if direct link is in outage and relay buffer has  $0 < m < D$  packets).
- If #decoding relays  $< \beta$ , packet is erroneously decoded at secondary destination, and will be retransmitted.
- When s-sd link is in outage, and source transmits, each relay sends 1 bit to indicate whether it is a decoding relay or not.
- If PU is active, secondary source is idle. SU relays forward the packets using relevant BF weights.
- Feedback: destination sends a feedback to indicate the status of decodability.

## CS Principles

- Applied when the relays send their states of being decoding relays or not.
- Instead of sending  $N$  orthogonal signals to source to describe state of each relay, CS techniques used to reduce the #samples.
- Received signal  $\mathbf{r}_j, j \in \{s, sd\}$ :

$$\mathbf{r}_j = \mathbf{A}\mathbf{H}_j^H \mathbf{x} + \mathbf{Z}_j,$$

$\mathbf{A} : [\mathbf{a}_1 \ \mathbf{a}_2 \ \cdots \ \mathbf{a}_N]$ ,  $N$ : Total number of relays.

- Need only  $h_{k,j} \times k$  without estimating  $h_{k,j}$ ; # Non-zero entries in  $\mathbf{H}_j^H \mathbf{x} = K$ .
- Random matrix  $\mathbf{A}$  is known to SU source and destination.

## Beamformer design

- Weights  $\mathbf{g}_a \in \mathbb{C}^K$ : active PU
- **Active PU**: Relays transmit one of the packets. Optimal weight vector  $\mathbf{g}_a$ :

$$\max_{\mathbf{g}_a} |\mathbf{g}_a^H \mathbf{h}_{sd}^{(\Omega)}|^2, \quad \text{s.t.} \quad |\mathbf{g}_a^H \mathbf{h}_{pd}^{(\Omega)}| = 0,$$

$$\mathbf{h}_{sd}^{(\Omega)} = [h_{1,sd}, \dots, h_{K,sd}], \quad \mathbf{h}_{pd}^{(\Omega)} = [h_{1,pd}, \dots, h_{K,pd}], \quad \Omega : \text{Set of chosen relays.}$$

## Performance Analysis

- Outage analysis, queuing analysis, PU throughput analysis, SU throughput maximization...

Paper: **Wideband Millimeter-Wave Propagation Measurements and Channel Models for Future Wireless Communication System Design.**

-T. S. Rappaport, G. R. MacCartney, Jr., M. K. Samimi, S. Sun.

Aim:

- Experimental measurements & empirically-based propagation channel models for 28, 38, 60, 73 GHz mmWave bands.
- Analysis of more than 15,000 power delay profiles in urban macrocells and microcells.

Specs:

- Wideband sliding correlator - ns-scale temporal resolution.
- Measurements for each frequency band - numerous large-scale path loss meas. for multiple AOA & AOD orientations of Tx and Rx.



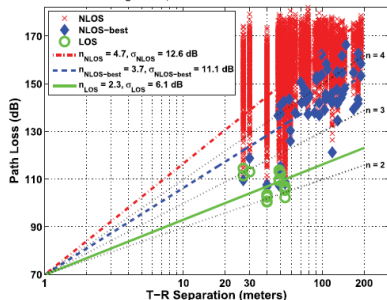
## Directional Path Loss Models

$$PL(d)[\text{dB}] = PL(d_0) + 10\bar{n} \log_{10} \left( \frac{d}{d_0} \right) + X_0, d \geq d_0, (d_0 = 1). \quad (1)$$

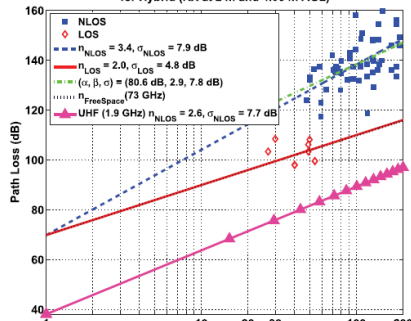
## Omnidirectional Path Loss Models

$$PL_{i,j}[\text{dB}] = Pt_{i,j} - 10 \log_{10} \left[ \sum_z \sum_y \sum_x \sum_w Pr_{i,j}(\theta_{r_w}, \phi_{r_x}, \theta_{t_y}, \phi_{t_z}) [\text{mW}] \right] \quad (2)$$

73 GHz Directional Path loss vs. Distance in Manhattan with RX Height: 2 m & 4.06 m  
Using 27 dBi, 7° 3dB BW TX and RX Antennas



73 GHz Omnidirectional PL Model 1 m - Manhattan  
for Hybrid (RX at 2 m and 4.06 m AGL)



## Beam Combining

- ↓ in PLEs for both coherent and non-coherent beam combining- ↑ coverage distance.

$$P_{\text{coh}} = \left( \sum_{i=1}^N \sqrt{P_i} \right)^2$$

$$P_{\text{non-coh}} = \sum_{i=1}^N P_i$$

$$d_{\text{beam}} = [d_{1\text{-beam}}]^{\text{DEE}}$$

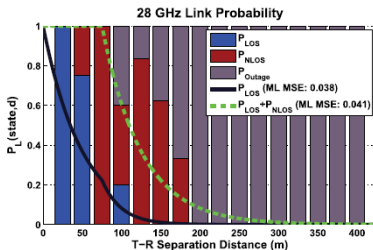
NLOS Directional Beam Combining Path Loss Models ( $d_0 = 1$ m)										
Frequency	TX Height (m)	RX Height (m)	TX / RX Antenna HPBW							
28 GHz (Man.)	7; 17	1.5	10.9° / 10.9°	PLE (Over all angles) = 4.556						
				Coherent			Non-Coherent			
				Beams	PLE	$\sigma$ [dB]	DEE	PLE	$\sigma$ [dB]	DEE
				1	3.812	9.1	-	3.812	9.1	-
				2	3.548	9.1	1.074	3.692	9.2	1.033
				3	3.406	9.2	1.119	3.631	9.2	1.050
				4	3.307	9.2	1.153	3.591	9.2	1.062
38 GHz	8; 23; 36	1.5	7.8° / 7.8°	PLE (Over all angles) = 3.295						
				Coherent			Non-Coherent			
				Beams	PLE	$\sigma$ [dB]	DEE	PLE	$\sigma$ [dB]	DEE
				1	2.801	12.2	-	2.801	12.2	-
				2	2.653	10.9	1.056	2.756	11.5	1.016
				3	2.579	10.6	1.086	2.741	11.4	1.022
				4	2.531	10.3	1.107	2.731	11.3	1.026
			PLE (Over all angles) = 2.826							

## Outage Studies

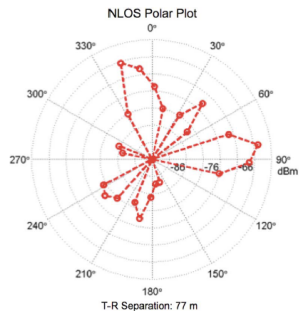
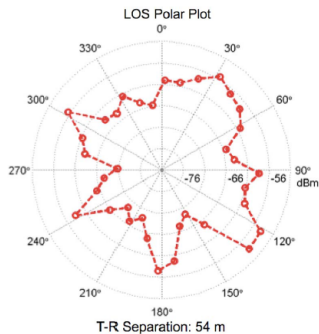
$$p_{\text{outage}}(d) = \max(0, 1 - e^{-a_{\text{out}}d + b_{\text{out}}}),$$

$$p_{\text{LOS}}(d) = (1 - p_{\text{outage}}(d))e^{-a_{\text{out}}d},$$

$$p_{\text{NLOS}}(d) = 1 - p_{\text{outage}}(d) - p_{\text{LOS}}(d).$$



## Spatial Properties



- RMSE delay spreads, multipath effects, peer-to-peer and vehicular channel responses...

Paper: **Antenna Grouping Based Feedback Compression for FDD-Based Massive MIMO Systems.**

-B. Lee, J. Choi, J.-Y. Seol, D. J. Love, B. Shim.

Aim:

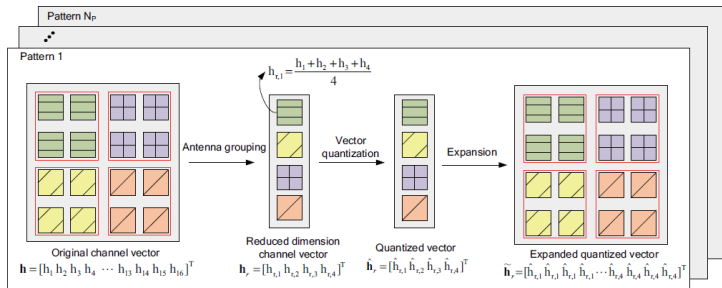
- Reduce the CSI feedback overhead from user-terminal to base station.
- Antenna grouping- to map multiple correlated antenna elements into one representative value.
- Lower quantization error than conventional vector quantization (channel-statistic based codeword).

Prelims:

- Multiuser multiple-input, single-output downlink channel
- Assume spatially and temporally correlated block-fading channels - channel vector  $\mathbf{h}_k$  follows first-order Gauss-Markov model.
- Data model:  $\mathbf{y}[n] = \mathbf{H}\mathbf{x}[n] + \mathbf{z}[n]$ ;  $\mathbf{y}[n] \in \mathbb{C}^{K \times 1}$ ,  $\mathbf{H}[n] \in \mathbb{C}^{K \times N_t}$
- To feedback CSI, user quantizes  $\mathbf{h}_k$ , chooses codeword from pre-defined codebook set.

## Antenna Grouping Based Feedback Reduction

- Reduce dimension of  $\mathbf{h}_k$  from  $N_t$  to  $N_g$  - grouping multiple correlated antennas.
- Feedback: header- group pattern, payload- index of quantized  $\mathbf{h}_k$ .
- Group pattern generation: AGB is sensitive to choice of antenna pattern.
- Either group antennas spatially, or use Grassmannian subspace packing. [D.J. Love et al. "Grassmannian beamforming..", 2003].
- Grassmannian approach has substantial gain over randomly selecting patterns.



## Some other papers

- Quickest Wideband Spectrum Sensing Over Correlated Channels- A. Tajer and J. Heydari.
- Energy-Efficient Resource Management in OFDM-Based Cognitive Radio Networks Under Channel Uncertainty - S. Wang, W. Shi, and C. Wang.
- Secrecy Performance Analysis for SIMO Simultaneous Wireless Information and Power Transfer Systems - G. Pan, C. Tang, T. Li, and Y. Chen.